

# Combinatorial and Algorithmic Aspects of Hyperbolic Polynomials

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## Abstract

Let  $p(x_1, \dots, x_n) = \sum_{(r_1, \dots, r_n) \in I_{n,n}} a_{(r_1, \dots, r_n)} \prod_{1 \leq i \leq n} x_i^{r_i}$  be homogeneous polynomial of degree  $n$  in  $n$  real variables with integer nonnegative coefficients. The support of such polynomial  $p(x_1, \dots, x_n)$  is defined as  $\text{supp}(p) = \{(r_1, \dots, r_n) \in I_{n,n} : a_{(r_1, \dots, r_n)} \neq 0\}$ . The convex hull  $CO(\text{supp}(p))$  of  $\text{supp}(p)$  is called the Newton polytope of  $p$ . We study the following decision problems, which are far-reaching generalizations of the classical perfect matching problem:

- **Problem 1**. Consider a homogeneous polynomial  $p(x_1, \dots, x_n)$  of degree  $n$  in  $n$  real variables with nonnegative integer coefficients given as a black box (oracle). *Is it true that  $(1, 1, \dots, 1) \in \text{supp}(p)$ ?*
- **Problem 2**. Consider a homogeneous polynomial  $p(x_1, \dots, x_n)$  of degree  $n$  in  $n$  real variables with nonnegative integer coefficients given as a black box (oracle). *Is it true that  $(1, 1, \dots, 1) \in CO(\text{supp}(p))$ ?*

We prove that for hyperbolic polynomials these two problems are equivalent and can be solved by deterministic polynomial-time oracle algorithms. This result is based on a "hyperbolic" generalization of the Rado theorem. We also present combinatorial and algebraic applications of this "hyperbolic" generalization of the Rado theorem (prove that the support  $\text{supp}(p)$  of  $P$ -hyperbolic polynomial  $p$  is an intersection of some Integral Polymatroid with the hyperplane  $\{(r_1, \dots, r_n) : \sum_{1 \leq i \leq n} r_i = n\}$ ) and pose some open problems.

## 1 Introduction and motivating examples

### The layout of the paper :

*We introduce the main topics and motivations in Section 1. In Section 1.1 we present a naive algorithm to solve Problem 1 in the general case. We show in Appendix D that this algorithm is, in a sense, optimal.*

*(Incidentally (or not), the situation here is very similar with the optimality of the square root in the famous quantum Grover's search algorithm.)*

*In Section 1.2 we remind the basic properties of hyperbolic polynomials used in this paper.*

*In Section 2 we state a hyperbolic analogue of the Rado theorem (Theorem 2.2), which is the main mathematical result of the paper. Theorem 2.2 sheds more light on the algebraic-geometric nature of such fundamental combinatorial results as Hall's and Rado's theorems.*

*In Section 2.1 we define and study doubly-stochastic polynomials (an useful generalization of standard doubly-stochastic matrices). We also state there a hyperbolic analogue of the van der Waerden conjecture.*

In Section 3 we introduce and analyse the ellipsoid algorithm which solves Problem 1 and Problem 2 on the class of  $S$ -hyperbolic polynomials . The essence of the results in Section 3 is that once Hall's like conditions (the exponential number of them) are proved to be necessary and sufficient , they can be checked by a polynomial time deterministic oracle algorithms . The algorithm , which we use , is based not on the linear programming but on some rather nonlinear convex programs similar to considered in [23] , [21] , [22].

In section 4 we introduce and analyse another algorithm , which is a "polynomial" generalization of the Sinkhorn Scaling .

In Section 5 we use Theorem 2.2 to get a more refine (polymatroidal) properties of the supports of  $P$ -hyperbolic polynomials and explain how our results generalize the main result from [7].

In Section 6 we pose some open problems and share our enthusiasm about the topic .

The proofs of the main results are presented in Appendices **A,B,C,D** .

Let  $p(x_1, \dots, x_n) = \sum_{(r_1, \dots, r_n) \in I_{n,n}} a_{(r_1, \dots, r_n)} \prod_{1 \leq i \leq n} x_i^{r_i}$  be homogeneous polynomial of degree  $n$  in  $n$  real variables. Here  $I_{k,n}$  stands for the set of vectors  $r = (r_1, \dots, r_k)$  with nonnegative integer components and  $\sum_{1 \leq i \leq k} r_i = n$ . In this paper we primarily study homogeneous polynomials with nonnegative integer coefficients .

**Definition 1.1:** The support of the polynomial  $p(x_1, \dots, x_n)$  as above is defined as  $\text{supp}(p) = \{(r_1, \dots, r_n) \in I_{n,n} : a_{(r_1, \dots, r_n)} \neq 0\}$  . The convex hull  $CO(\text{supp}(p))$  of  $\text{supp}(p)$  is called the Newton polytope of  $p$  . ■

We will study the following decision problems :

- **Problem 1** . Consider a homogeneous polynomial  $p(x_1, \dots, x_n)$  of degree  $n$  in  $n$  real variables with nonnegative integer coefficients given as a black box (oracle) . Is it true that  $(1, 1, \dots, 1) \in \text{supp}(p)$  ?  
An equivalent question is : Is it true that  $\frac{\partial^n}{\partial x_1 \dots \partial x_n} p(x_1, \dots, x_n) \neq 0$  ?
- **Problem 2** . Consider a homogeneous polynomial  $p(x_1, \dots, x_n)$  of degree  $n$  in  $n$  real variables with nonnegative integer coefficients given as a black box (oracle) . Is it true that  $(1, 1, \dots, 1) \in CO(\text{supp}(p))$  ?

Our goal is solve these decision problems using deterministic polynomial-time oracle algorithms , i.e. algorithms which evaluate the given polynomial  $p(\cdot)$  at a number of rational vectors  $(q_1, \dots, q_n)$  which is polynomial in  $n$  and  $\log(p(1, 1, \dots, 1))$ ; these rational vectors  $(q_1, \dots, q_n)$  are supposed to have bit-wise complexity which is polynomial in  $n$  and  $\log(p(1, 1, \dots, 1))$  ; and the additional auxiliary arithmetic computations also take a polynomial number of steps in  $n$  and  $\log(p(1, 1, \dots, 1))$  .

The next example explains some (well known ) origins of the both problems .

**Example 1.2:** Consider first the following homogeneous polynomial from [28] :  $p(x_1, \dots, x_n) = \text{tr}((D(x)A)^n)$  , where  $D(x)$  is a  $n \times n$  diagonal matrix  $\text{Diag}(x_1, \dots, x_n)$  ; and  $A$  is  $n \times n$  matrix with  $(0, 1)$  entries , i.e.  $A$  is an adjacency matrix of some directed graph  $\Gamma$  . Clearly , this polynomial  $p(x_1, \dots, x_n)$  has nonnegative integer coefficients . It was proved in [28] that  $\frac{1}{n} \frac{\partial^n}{\partial x_1 \dots \partial x_n} \text{tr}((D(x)A)^n)$  is equal to the number of Hamiltonian circuits in the graph  $\Gamma$  . Notice that the polynomial  $\text{tr}(D(x)A)^n$  can be evaluated in  $O(n^3 \log(n))$  arithmetic operations and  $(1, 1, \dots, 1) \in \text{supp}(p)$  iff there exists a Hamiltonian circuit in the graph  $\Gamma$  . Also  $\log(p(1, 1, \dots, 1)) \leq n \log(n)$ . Therefore , unless  $P = NP$  , there is no hope to design deterministic polynomial-time oracle algorithm solving Problem 1 in this case . (The author is indebted to A.Barvinok for pointing out this polynomial . )

Consider , with the same adjacency matrix  $A$  , another homogeneous polynomial

$Mul(x_1, \dots, x_n) = \prod_{1 \leq i \leq n} \sum_{1 \leq j \leq n} A(i, j)x_j$  . Then

$$\frac{\partial^n}{\partial x_1 \dots \partial x_n} Mul(x_1, \dots, x_n) = \text{Per}(A) .$$

Therefore for the multilinear polynomial  $Mul(x_1, \dots, x_n)$  Problem 1 is a "black box" analogue of checking the existence of the perfect bipartite matching .

Next consider the following class of determinantal polynomials :

$$q(x_1, \dots, x_n) = \det\left(\sum_{1 \leq i \leq n} A_i x_i\right),$$

where  $\mathbf{A} = (A_1, \dots, A_n)$  is a  $n$ -tuple of positive semidefinite  $n \times n$  hermitian matrices , i.e.  $A_i \succeq 0$  , with integer entries . Recall that the mixed discriminant

$$D(\mathbf{A}) = \frac{\partial^n}{\partial \alpha_1 \dots \partial \alpha_n} \det\left(\sum_{1 \leq i \leq n} A_i x_i\right).$$

(If the matrices  $A_i$  above are diagonal , i.e.  $A_i = \text{Diag}(b_{i,1}, \dots, b_{i,n})$ ,  $1 \leq i \leq n$  , then the mixed discriminant is reduced to the permanent :  $D(\mathbf{A}) = \text{Per}(B)$ ,  $B = \{b_{i,j}; 1 \leq i, j \leq n\}$  ).

It is well known (see , for instance , [22] ) that a determinantal polynomial  $q(\cdot)$  can be represented as

$$q(x_1, \dots, x_n) = \sum_{r \in I_{n,n}} \prod_{1 \leq i \leq n} x_i^{r_i} D(\mathbf{A}_r) \frac{1}{\prod_{1 \leq i \leq n} r_i!}, \quad (1)$$

where a  $n$ -tuple  $\mathbf{A}_r$  of square matrices consists of  $r_i$  copies of  $A_i$ ,  $1 \leq i \leq k$  . One of the equivalent formulations [34] of the classical Rado theorem states that  $D(\mathbf{A}_{(1,1,\dots,1)}) > 0$  iff

$$\text{Rank}\left(\sum_{i \in S} A_i\right) \geq |S| \text{ for all } S \subset \{1, 2, \dots, n\} \quad (2)$$

(The diagonal case is the famous Hall's theorem on the perfect bipartite matchings .)

The Rado theorem is just a particular case of famous Edmonds theorem on the rank of intersection of two matroids . Therefore , given a  $n$ -tuple  $\mathbf{A} = (A_1, \dots, A_n)$  of positive semidefinite  $n \times n$  hermitian matrices , one can decide in deterministic polynomial time if  $D(\mathbf{A}) > 0$  . We

will explain below that this decision problem can be solved by a deterministic polynomial-time oracle algorithm . I.e. we only use some values of  $\det(\sum_{1 \leq i \leq n} A_i x_i)$  without reconstructing the actual tuple  $\mathbf{A} = (A_1, \dots, A_n)$  .

**The natural question , in our opinion , is which algebraic-geometric properties make the class of determinantal polynomials "easy" and the class of Barvinok's polynomials  $\text{tr}(D(x)A)^n$  "hard" . This paper suggests one answer to the question.**

One important corollary of the Rado conditions (2) is that

$$\text{supp}(q) = CO(\text{supp}(q)) \cap I_{n,n}. \quad (3)$$

I.e. if integer vectors  $r, r(1), r(2), \dots, r(k) \in I(n, n)$  and

$$r = \sum_{1 \leq i \leq k} a(i)r(i), a(i) \geq 0, 1 \leq i \leq k; \sum_{1 \leq i \leq k} a(i),$$

and  $D(\mathbf{A}_{r(i)}) > 0, 1 \leq i \leq k$  then also  $D(\mathbf{A}_r) > 0$  . Notice that in this case Problem 1 and Problem 2 are equivalent .

We can rewrite Rado conditions (2) as follows :

$$\max_{r \in \text{supp}(q)} \sum_{i \in S} r_i \geq |S| \text{ for all } S \subset \{1, 2, \dots, n\} \quad (4)$$

Putting things together we get the following Fact .

**Fact 1.3:** The following properties of determinantal polynomial  $q(x_1, \dots, x_n) = \det(\sum_{1 \leq i \leq n} A_i x_i)$  with  $n \times n$  hermitian matrices  $A_i \succeq 0, 1 \leq i \leq n$  are equivalent .

1.  $(1, 1, \dots, 1) \notin \text{supp}(q)$ .
2.  $(1, 1, \dots, 1) \notin CO(\text{supp}(q))$ .
3. There exists nonempty  $S \subset \{1, 2, \dots, n\}$  such that

$$\sum_{1 \leq i \leq n} r_i s_i < \sum_{1 \leq i \leq n} s_i = |S| \text{ for all } (r_1, \dots, r_n) \in \text{supp}(q), \quad (5)$$

where  $(s_1, \dots, s_n)$  is a characteristic function of the subset  $S$  , i.e.  $s_i = 1$  if  $i \in S$  , and  $s_i = 0$  otherwise .

*Notice that if (5) holds then the distance  $\text{dist}(e, CO(\text{supp}(q)))$  from the vector  $e = (1, \dots, 1)$  to the Newton polytope  $CO(\text{supp}(q))$  is at least  $\sqrt{\frac{n}{|S|(n-|S|)}} \geq \frac{2}{\sqrt{n}}$  .*

■

We will show that for any class of polynomials satisfying Fact 1.3 there exists a deterministic polynomial-time oracle algorithm solving both Problem 1 and Problem 2 , which are , of course , equivalent in this case . Our algorithm is based on the reduction to some convex programming problem and the consequent use of the Ellipsoids method .

The next fact about determinantal polynomials , namely their hyperbolicity , is "responsible" for Fact 1.3 .

**Fact 1.4:** Consider a determinantal polynomial  $q((x_1, \dots, x_n) = \det(\sum_{1 \leq i \leq n} A_i x_i)$  with  $A_i \succeq 0, 1 \leq i \leq n$  . Assume that  $q$  is not identically zero , i.e. that  $B =: \sum_{1 \leq i \leq n} A_i \succ 0$  (the sum is strictly positive definite ) . For a real vector  $(x_1, \dots, x_n) \in R^n$  consider the following polynomial equation of degree  $n$  in one variable :

$$P(t) = q(x_1 - t, x_2 - t, \dots, x_n - t) = \det\left(\sum_{1 \leq i \leq n} A_i x_i - t \sum_{1 \leq i \leq n} A_i\right) = 0. \quad (6)$$

Equation (6) has  $n$  real roots counting the multiplicities ; if the real vector  $(x_1, \dots, x_n) \in R^n$  has nonnegative entries then all roots of (6) are nonnegative real numbers . ■

The main result of this paper that this hyperbolicity , which we will describe formally in Section 1.1 , is sufficient for Fact 1.3 ; i.e. Fact 1.4 implies Fact 1.3 . ■

## 1.1 "Naive" algorithms

One possible "naive" algorithm to solve Problem 1 is just to compute  $\frac{\partial^n}{\partial x_1 \dots \partial x_n} p(x_1, \dots, x_n)$  . Recall that the number of coefficients of a homogeneous polynomial of degree  $n$  in  $n$  real variables is equal to  $\frac{(2n-1)!}{n!(n-1)!} \approx 2^{2n}$  . We can compute all the coefficients of  $p(x_1, \dots, x_n)$  via standard multidimensional interpolation , but this interpolation will need  $\frac{(2n-1)!}{n!(n-1)!} \approx 2^{2n}$  oracle calls . There is an algorithm which computes  $\frac{\partial^n}{\partial x_1 \dots \partial x_n} p(x_1, \dots, x_n)$  using only  $2^{n-1}$  oracle calls :

$$\frac{\partial^n}{\partial x_1 \dots \partial x_n} p(x_1, \dots, x_n) = 2^{-n+1} \sum_{b_i \in \{-1, +1\}, 2 \leq i \leq n} p(1, b_2, \dots, b_n) \prod_{2 \leq i \leq n} b_i. \quad (7)$$

This formula is , in a sense , optimal . I.e. there exists a nearly matching lower bound . The corresponding result and connections to computations of the permanent are presented in Appendix D .

We will explain below that if  $p$  is a homogeneous polynomial of degree  $n$  in  $n$  real variables with nonnegative integer coefficients then  $(1, 1, \dots, 1) \in CO(\text{supp}(p))$  iff  $p(x_1, \dots, x_n) \geq \prod_{1 \leq i \leq n} x_i$  for all vectors  $(x_1, \dots, x_n)$  with positive real coordinates . Therefore Problem 2 is equivalent to checking if the polynomial  $P(y_1, \dots, y_n) = p(1 + y_1^2, \dots, 1 + y_n^2) - \prod_{1 \leq i \leq n} 1 + y_i^2$  is nonnegative on  $R^n$  .

## 1.2 Hyperbolic polynomials

The following concept of hyperbolic polynomials was originated in the theory of partial differential equations [18], [9], [10].

A homogeneous polynomial  $p(x)$ ,  $x \in R^m$  of degree  $n$  in  $m$  real variables is called hyperbolic in the direction  $e \in R^m$  (or  $e$ -hyperbolic) if for any  $x \in R^m$  the polynomial  $p(x - \lambda e)$  in the one variable  $\lambda$  has exactly  $n$  real roots counting their multiplicities. We assume in this paper that  $p(e) > 0$ . Denote an ordered vector of roots of  $p(x - \lambda e)$  as  $\lambda(x) = (\lambda_1(x) \geq \lambda_2(x) \geq \dots \lambda_n(x))$ . It is well known that the product of roots is equal to  $p(x)$ . Call  $x \in R^m$   $e$ -positive ( $e$ -nonnegative) if  $\lambda_n(x) > 0$  ( $\lambda_n(x) \geq 0$ ). The fundamental result [18] in the theory of hyperbolic polynomials states that the set of  $e$ -nonnegative vectors is a closed convex cone. A  $k$ -tuple of vectors  $(x_1, \dots, x_k)$  is called  $e$ -positive ( $e$ -nonnegative) if  $x_i$ ,  $1 \leq i \leq k$  are  $e$ -positive ( $e$ -nonnegative). We denote the closed convex cone of  $e$ -nonnegative vectors as  $N_e(p)$ , and the open convex cone of  $e$ -positive vectors as  $C_e(p)$ .

*Recent interest in the hyperbolic polynomials got sparked by the discovery [12], [11] that  $\log(p(x))$  is a self-concordant barrier for the opened convex cone  $C_e(p)$  and therefore the powerful machinery of interior-point methods can be applied. It is an important open problem whether this cone  $C_e(p)$  has a semi-definite representation.*

It has been shown in [18] (see also [26]) that an  $e$ -hyperbolic polynomial  $p$  is also  $d$ -hyperbolic for all  $e$ -positive vectors  $d \in C_e(p)$ .

Let us fix  $n$  real vectors  $x_i \in R^m$ ,  $1 \leq i \leq n$  and define the following homogeneous polynomial:

$$P_{x_1, \dots, x_n}(\alpha_1, \dots, \alpha_n) = p\left(\sum_{1 \leq i \leq n} \alpha_i x_i\right) \quad (8)$$

Following [26], we define the  $p$ -mixed form of an  $n$ -vector tuple  $\mathbf{X} = (x_1, \dots, x_n)$  as

$$M_p(\mathbf{X}) =: M_p(x_1, \dots, x_n) = \frac{\partial^n}{\partial \alpha_1 \dots \partial \alpha_n} p\left(\sum_{1 \leq i \leq n} \alpha_i x_i\right) \quad (9)$$

Equivalently, the  $p$ -mixed form  $M_p(x_1, \dots, x_n)$  can be defined by the polarization (see [26]):

$$M_p(x_1, \dots, x_n) = 2^{-n} \sum_{b_i \in \{-1, +1\}, 1 \leq i \leq n} p\left(\sum_{1 \leq i \leq n} b_i x_i\right) \prod_{1 \leq i \leq n} b_i \quad (10)$$

Associate with any vector  $r = (r_1, \dots, r_n) \in I_{n,n}$  an  $n$ -tuple of  $m$ -dimensional vectors  $\mathbf{X}_r$  consisting of  $r_i$  copies of  $x_i$  ( $1 \leq i \leq n$ ). It follows from the Taylor's formula that

$$P_{x_1, \dots, x_n}(\alpha_1, \dots, \alpha_n) = \sum_{r \in I_{n,n}} \prod_{1 \leq i \leq n} \alpha_i^{r_i} M_p(\mathbf{X}_r) \frac{1}{\prod_{1 \leq i \leq n} r_i!} \quad (11)$$

For an  $e$ -nonnegative tuple  $\mathbf{X} = (x_1, \dots, x_n)$ , define its capacity as:

$$Cap(\mathbf{X}) = \inf_{\alpha_i > 0, \prod_{1 \leq i \leq n} \alpha_i = 1} P_{x_1, \dots, x_n}(\alpha_1, \dots, \alpha_n) \quad (12)$$

Probably the best known example of a hyperbolic polynomial comes from the hyperbolic geometry :

$$P(\alpha_0, \dots, \alpha_k) = \alpha_0^2 - \sum_{1 \leq i \leq k} \alpha_i^2 \quad (13)$$

This polynomial is hyperbolic in the direction  $(1, 0, 0, \dots, 0)$ . Another "popular" hyperbolic polynomial is  $\det(X)$  restricted on a linear real space of hermitian  $n \times n$  matrices . In this case mixed forms are just mixed discriminants , hyperbolic direction is the identity matrix  $I$  , the corresponding closed convex cone of  $I$ -nonnegative vectors coincides with a closed convex cone of positive semidefinite matrices .

Less known , but very interesting , hyperbolic polynomial is the Moore determinant  $Det_{(M)}(Y)$  restricted on a linear real space of hermitian quaternionic  $n \times n$  matrices . The Moore determinant is , essentially , the Pfaffian (see the corresponding definitions and the theory in a very readable paper [38] ) .

We use in this paper the following class of hyperbolic in the direction  $(1, 1, \dots, 1)$  polynomials of degree  $k$  :

$Q(\alpha_1, \dots, \alpha_k) = M_p(\sum_{1 \leq i \leq k} \alpha_i x_i, \dots, \sum_{1 \leq i \leq k} \alpha_i x_i, x_{k+1}, \dots, x_n)$ , where  $p$  is a  $e$ -hyperbolic polynomial of degree  $n > k$  ,  $(x_1, \dots, x_n)$  is  $e$ -nonnegative tuple , and the  $p$ -mixed form  $M_p(\sum_{1 \leq i \leq k} x_i, \dots, \sum_{1 \leq i \leq k} x_i, x_{k+1}, \dots, x_n) > 0$ .

We make a substantial use of the following very recent result [27] , which is a rather direct corollary of [1] , [37] .

**Theorem 1.5:** *Consider a homogeneous polynomial  $p(y_1, y_2, y_3)$  of degree  $n$  in 3 real variables which is hyperbolic in the direction  $(0, 0, 1)$ . Assume that  $p(0, 0, 1) = 1$  . Then there exists two  $n \times n$  real symmetric matrices  $A, B$  such that*

$$p(y_1, y_2, y_3) = \det(y_1 A + y_2 B + y_3 I).$$

It has been shown in [19] that most of known facts , and some opened problems as well , about hyperbolic polynomials follow from Theorem 1.5 .

## 2 A hyperbolic analogue of the Rado theorem

**Definition 2.1:** Consider a homogeneous polynomial  $p(x), x \in R^m$  of degree  $n$  in  $m$  real variables which is hyperbolic in the direction  $e$ . Denote an ordered vector of roots of  $p(x - \lambda e)$  as  $\lambda(x) = (\lambda_1(x) \geq \lambda_2(x) \geq \dots \lambda_n(x))$  . We define the  $p$ -rank of  $x \in R^m$  in direction  $e$  as  $Rank_p(x) = |\{i : \lambda_i(x) \neq 0\}|$ . It follows from Theorem 1.5 that the  $p$ -rank of  $x \in R^m$  in any direction  $d \in C_e$  is equal to the  $p$ -rank of  $x \in R^m$  in direction  $e$  , which we call the  $p$ -rank of  $x \in R^m$  . ■

Consider the following polynomial in one variable  $D(t) = p(td + x) = \sum_{0 \leq i \leq n} c_i t^i$ . It follows from the identity (11) that

$$\begin{aligned}
c_n &= M_p(d, \dots, d)(n!)^{-1} = p(d), \\
c_{n-1} &= M_p(x, d, \dots, d)(1!(n-1)!)^{-1}, \dots, \\
c_0 &= M_p(x, \dots, x)(n!)^{-1} = p(x).
\end{aligned} \tag{14}$$

Let  $(\lambda_1^{(d)}(x) \geq \lambda_2^{(d)}(x) \geq \dots \geq \lambda_n^{(d)}(x))$  be the (real) roots of  $x$  in the  $e$ -positive direction  $d$ , i.e. the roots of the equation  $p(td - x) = 0$ . Define (canonical symmetric functions) :

$$S_{k,d}(x) = \sum_{1 \leq i_1 < i_2 < \dots < i_k \leq n} \lambda_{i_1}(x) \lambda_{i_2}(x) \dots \lambda_{i_k}(x).$$

Then  $S_{k,d}(x) = \frac{c_{n-k}}{c_n}$ . Clearly if  $x$  is  $e$ -nonnegative then for any  $e$ -positive vector  $d$  the  $p$ -rank  $\text{Rank}_p(x) = \max\{k : S_{k,d}(x) > 0\}$ . The next theorem, which we prove in Appendix A, is the main mathematical result of this paper. Our main tool is Theorem 1.5, which facilitates a rather easy induction. We also use a particularly easy case of the Rado theorem (see Remark A.5 for the details).

**Theorem 2.2:** Consider a homogeneous polynomial  $p(x), x \in R^m$  of degree  $n$  in  $m$  real variables which is hyperbolic in the direction  $e$ ,  $p(e) > 0$ . Let  $\mathbf{X} = (x_1, \dots, x_n), x_i \in R^m$  be  $e$ -nonnegative  $n$ -tuple of  $m$ -dimensional vectors, i.e.  $x_i, 1 \leq i \leq n$  are  $e$ -nonnegative.

Then the  $p$ -mixed form  $M_p(\mathbf{X}) =: M_p(x_1, \dots, x_n)$  is positive iff the following generalized Rado conditions hold :

$$\text{Rank}_p\left(\sum_{i \in S} x_i\right) \geq |S| \text{ for all } S \subset \{1, 2, \dots, n\}. \tag{15}$$

**Definition 2.3:** Call a homogeneous polynomial  $p(\alpha), \alpha \in R^n$  of degree  $n$  in  $n$  real variables  $P$ -hyperbolic if it is hyperbolic in direction  $e = (1, 1, \dots, 1)$  (vector of all ones),  $p(e) > 0$  and all the canonical orts  $e_i, 1 \leq i \leq n$  (rows of the identity matrix  $I$ ) are  $e$ -nonnegative. In other words, a homogeneous polynomial  $p(\alpha), \alpha \in R^n$  of degree  $n$  in  $n$  real variables is  $P$ -hyperbolic if it is  $e$ -hyperbolic and its closed cone of  $e$ -nonnegative vectors contains the nonnegative orthant  $R_+^n = \{(x_1, \dots, x_n) : x_i \geq 0, 1 \leq i \leq n\}$ . It follows from [26] that the coefficients of  $P$ -hyperbolic polynomials are nonnegative real numbers.

(Notice that the class of  $P$ -hyperbolic polynomials coincides with the class of polynomials  $P_{x_1, \dots, x_n}(\alpha_1, \dots, \alpha_n) = p(\sum_{1 \leq i \leq n} \alpha_i x_i)$ , where  $p$  is  $e$ -hyperbolic polynomial of degree  $n$  in  $m$  real variables, a  $n$ -tuple  $(x_1, \dots, x_n)$  of  $m$ -dimensional real vectors is  $e$ -nonnegative and  $\sum_{1 \leq i \leq n} x_i$  is  $e$ -positive.)

Call a homogeneous polynomial  $q(\alpha), \alpha \in R^n$  of degree  $n$  in  $n$  real variables with nonnegative coefficients  $S$ -hyperbolic if there exists a  $P$ -hyperbolic polynomial  $p$  such that  $\text{supp}(p) = \text{supp}(q)$ . ■

(One natural class of  $S$ -hyperbolic polynomials, not all of them  $P$ -hyperbolic, is  $\text{Vol}(\alpha_1 X_1 + \dots + \alpha_n X_n)$ , where  $X_i$  are convex compact subsets of  $R^n$ . See the corresponding not  $P$ -hyperbolic example in [26].)



**Corollary 2.4:** Let  $q(\alpha), \alpha \in R^n$  be  $S$ -hyperbolic polynomial of degree  $n$  .  
Then  $CO(\text{supp}(q)) \cap I_{n,n} = \text{supp}(q)$  .

**Proof:** It is enough to prove the corollary for  $P$ -hyperbolic polynomials. I.e. suppose that  $q(\alpha_1, \dots, \alpha_n) = p(\sum_{1 \leq i \leq n} \alpha_i x_i)$  , where  $p$  is  $e$ -hyperbolic polynomial of degree  $n$  in  $m$  real variables , a  $n$ -tuple  $(x_1, \dots, x_n)$  of  $m$ -dimensional real vectors is  $e$ -nonnegative and  $\sum_{1 \leq i \leq n} x_i$  is  $e$ -positive . Then  $r = (r_1, r_2, \dots, r_n) \in \text{supp}(q)$  iff the  $p$ -mixed form  $M_p(\mathbf{X}_r) > 0$  , where the  $n$ -tuple  $\mathbf{X}_r$  consists of  $r_i$  copies of  $x_i, 1 \leq i \leq n$ . Let  $r^{(0)} = (r_1^{(0)}, \dots, r_n^{(0)}) \in CO(\text{supp}(q))$ . I.e. there exist  $r^{(j)} \in \text{supp}(q), 1 \leq j \leq n$  such that  $r^{(0)} = \sum_{1 \leq j \leq n} a_j r^{(j)}$  and  $a_j \geq 0, \sum_{1 \leq j \leq n} a_j = 1$  .

Let  $r^{(j)} = (r_1^{(j)}, \dots, r_n^{(j)}), 0 \leq j \leq n$  . As  $r^{(j)} \in \text{supp}(q), 1 \leq j \leq n$  thus  $M_p(\mathbf{X}_{r^{(j)}}) > 0, 1 \leq j \leq n$  . It follows from Theorem 2.2 (only if part ) that

$$\text{Rank}_p(\sum_{i \in S} x_i) \geq \sum_{i \in S} r_i^{(j)} \text{ for all } S \subset \{1, 2, \dots, n\}; 1 \leq j \leq n.$$

Therefore

$$\text{Rank}_p(\sum_{i \in S} x_i) \geq \sum_{i \in S} \sum_{1 \leq j \leq n} a_j r_i^{(j)} = \sum_{i \in S} r_i^{(j)}, S \subset \{1, 2, \dots, n\}.$$

Using the "if" part of Theorem 2.2 we get that  $M_p(\mathbf{X}_{r^{(0)}}) > 0$  and thus  $r^{(0)} \in \text{supp}(q)$  . ■

**Corollary 2.5:** Let  $q(x), x \in R^n$  be  $S$ -hyperbolic polynomial of degree  $n$  . Then the following conditions are equivalent

1.  $e \in CO(\text{supp}(q))$  .
2.  $e \in \text{supp}(q)$  , i.e.  $\frac{\partial^n}{\partial \alpha_1 \dots \partial \alpha_n} q(x) > 0$  .
3.  $\text{Cap}(p) =: \inf_{\alpha_i > 0, \prod_{1 \leq i \leq n} \alpha_i = 1} q(\alpha_1, \dots, \alpha_n) > 0$ .
4. For all  $\epsilon > 0$  there exists a vector  $(\alpha_1, \dots, \alpha_n)$  with positive entries such that the following inequality holds :

$$\sum_{1 \leq i \leq n} \left| \frac{\alpha_i \frac{\partial}{\partial \alpha_i} q(\alpha_1, \dots, \alpha_n)}{q(\alpha_1, \dots, \alpha_n)} - 1 \right|^2 \leq \epsilon. \quad (16)$$

5. There exists a vector  $(\alpha_1, \dots, \alpha_n)$  with positive entries such that the following inequality holds :

$$\sum_{1 \leq i \leq n} \left| \frac{\alpha_i \frac{\partial}{\partial \alpha_i} q(\alpha_1, \dots, \alpha_n)}{q(\alpha_1, \dots, \alpha_n)} - 1 \right|^2 \leq \frac{1}{n}. \quad (17)$$

6. For all subsets  $S \subset \{1, 2, \dots, n\}$  the following inequality holds :

$$\sum_{i \in S} r_i \geq |S| \text{ for all } (r_1, \dots, r_n) \in \text{supp}(q). \quad (18)$$

(We sketch a proof in Appendix C . )

The following result , which we prove in Appendix B , is a "polynomial" generalization of Lemma 4.2 in [20] .

**Proposition 2.6:** *The condition (17) implies the condition (18) for all homogeneous polynomial  $q(x), x \in R^n$  of degree  $n$  in  $n$  real variables with nonnegative coefficients .*

## 2.1 Doubly-stochastic polynomials

Inequalities (16), (17) above suggest the following definition .

**Definition 2.7:** A homogeneous polynomial  $q(x_1, \dots, x_n)$  of degree  $n$  in  $n$  variables is called doubly-stochastic if its coefficients are nonnegative real numbers and  $\frac{\partial}{\partial x_i} q(1, 1, \dots, 1) = 1$  for all  $1 \leq i \leq n$ . The doubly-stochastic defect of the polynomial  $q$  is defined as  $DS(q) = \sum_{1 \leq i \leq n} (\frac{\partial}{\partial x_i} q(1, 1, \dots, 1) - 1)^2$  ■

**Lemma 2.8:**

1. A homogeneous polynomial  $q(x_1, \dots, x_n)$  of degree  $n$  in  $n$  variables with nonnegative real coefficients is doubly-stochastic iff  $q(1, 1, \dots, 1) = 1$  and  $q(x_1, \dots, x_n) \geq \prod_{1 \leq i \leq n} x_i$  for all real vectors  $(x_1, \dots, x_n) \in R^n$  with positive coordinates (in other words if  $q(1, 1, \dots, 1) = 1$  and  $Cap(q) = 1$  ).
2. A homogeneous polynomial  $q(x_1, \dots, x_n)$  of degree  $n$  in  $n$  variables is  $P$ -hyperbolic and doubly-stochastic iff  $q(1, 1, \dots, 1) = 1$  and  $|q(z_1, \dots, z_n)| \geq \prod_{1 \leq i \leq n} Re(z_i)$  for all complex vectors  $(z_1, \dots, z_n) \in C^n$  with positive real parts .
3. If a sequence  $q_i$  of homogeneous polynomials of degree  $n$  in  $n$  variables with nonnegative real coefficients converges to a doubly-stochastic polynomial then  $\lim_{i \rightarrow \infty} Cap(q_i) = 1$ .
4. The capacity  $Cap(q)$  is a continuous functional (but not even Lipschitz) on a convex closed cone of homogeneous polynomials of degree  $n$  in  $n$  variables with nonnegative real coefficients .
5. If  $q$  is homogeneous polynomial of degree  $n$  in  $n$  variables with nonnegative real coefficients then  $Cap(q) > 0$  iff there exists a sequence  $X_j = (x_{1,j}, \dots, x_{n,j})$  of vectors with positive real coordinates such that

$$\sum_{1 \leq i \leq n} \left| \frac{x_i \frac{\partial}{\partial x_i} q(x_{1,j}, \dots, x_{n,j})}{q(x_{1,j}, \dots, x_{n,j})} - 1 \right|^2 \rightarrow 0$$

And in this case  $Cap(q) = \lim_{j \rightarrow \infty} \frac{q(X_j)}{\prod_{1 \leq i \leq n} x_{i,j}}$ .

**Example 2.9:** A multilinear polynomial

$q(x_1, \dots, x_n) = \prod_{1 \leq i \leq n} \sum_{1 \leq j \leq n} a(i, j) x_j$  is doubly-stochastic and  $P$ -hyperbolic iff the square matrix  $B = \{ \frac{a(i, j)}{\sum_{1 \leq k \leq n} a(i, k)} : 1 \leq i, j \leq n \}$  is doubly stochastic in the standard meaning . ■

The next theorem is another Corollary of Theorem 2.2 ; its main point is in introducing a hyperbolic analog of Van der Waerden conjecture .

**Theorem 2.10:**

1. Consider the set  $PHDS(n)$  of all  $P$ -hyperbolic doubly-stochastic homogeneous polynomials  $q(x_1, \dots, x_n)$  of degree  $n$  in  $n$  variables. The set  $PHDS(n)$  is a compact and the following inequality holds

$$\inf_{q \in PHDS(n)} \frac{\partial^n}{\partial x_1 \dots \partial x_n} q(x_1, \dots, x_n) = \min_{q \in PHDS(n)} \frac{\partial^n}{\partial x_1 \dots \partial x_n} q(x_1, \dots, x_n) =: VdW(n) > 0$$

2. For any  $P$ -hyperbolic homogeneous polynomials  $q(x_1, \dots, x_n)$  of degree  $n$  in  $n$  variables the following inequalities hold

$$VdW(n) \leq \frac{\frac{\partial^n}{\partial x_1 \dots \partial x_n} q(x_1, \dots, x_n)}{Cap(q)} \leq 1.$$

**Conjecture 2.11: Hyperbolic Van der Waerden conjecture**

$$VdW(n) = \frac{n!}{n^n}?$$

■

The next result , a direct corollary of Lemma 2.10 in [19] , is a generalization of Proposition 4.2 in [22].

**Lemma 2.12:** Let  $q$  be  $P$ -hyperbolic homogeneous polynomial of degree  $n$  in  $n$  variables . Then the following inequalities hold for all vectors  $(x_1, \dots, x_n)$  with positive real coordinates.

- 1.

$$Cap(q) \leq \frac{q(x_1, \dots, x_n)}{\prod_{1 \leq i \leq n} x_i} \prod_{1 \leq i \leq n} x_i \frac{\frac{\partial}{\partial x_i} q(x_1, \dots, x_n)}{q(x_1, \dots, x_n)} \quad (19)$$

2. If  $\log(\frac{q(x_1, \dots, x_n)}{\prod_{1 \leq i \leq n} x_i}) - \log(Cap(q)) \leq \frac{\epsilon}{10}$  with  $0 \leq \epsilon \leq 1$  then

$$\sum_{1 \leq i \leq n} \left| \frac{x_i \frac{\partial}{\partial x_i} q(x_1, \dots, x_n)}{q(x_1, \dots, x_n)} - 1 \right|^2 \leq \epsilon \quad (20)$$

**Example 2.13:** Consider the following homogeneous polynomial of degree  $n$  in  $n$  variables :  $p(x_1, \dots, x_n) = \sum_{1 \leq i \leq n} x_i^n$ . Then  $Cap(p) = n$  and

$$\begin{aligned} & \frac{p(x_1, \dots, x_n)}{\prod_{1 \leq i \leq n} x_i} \prod_{1 \leq i \leq n} x_i \frac{\frac{\partial}{\partial x_i} p(x_1, \dots, x_n)}{p(x_1, \dots, x_n)} = \\ & = n^n \left( \frac{\prod_{1 \leq i \leq n} x_i}{\sum_{1 \leq i \leq n} x_i^n} \right)^{n-1} \leq n = Cap(p). \end{aligned}$$

The moral of this example is that the inequality (19) does not hold for all homogeneous polynomials with nonnegative coefficients, this inequality is a nontrivial necessary condition for the  $P$ -hyperbolicity. It is interesting to notice that the inequality (19) implies the determinantal Hadamard inequality. ■

**Remark 2.14:** Let

$$p(x_1, \dots, x_n) = \sum_{(r_1, \dots, r_n) \in I_{n,n}} a_{(r_1, \dots, r_n)} \prod_{1 \leq i \leq n} x_i^{r_i}$$

be homogeneous polynomial of degree  $n$  in  $n$  real variables. Assume that its coefficients are nonnegative and sum to one, i.e. that  $p(1, 1, \dots, 1) = 1$ . Associate with this polynomial a random integer vector

$$Z_p = (z_1, \dots, z_n) \in I_{n,n} : Prob\{Z_p = (r_1, \dots, r_n)\} = a_{(r_1, \dots, r_n)}.$$

Then

$$E(Z_p) = \left( \frac{\partial}{\partial x_1} p(1, 1, \dots, 1), \dots, \frac{\partial}{\partial x_n} p(1, 1, \dots, 1) \right).$$

Therefore  $p$  is doubly-stochastic iff  $E(Z_p) = (1, 1, \dots, 1)$ ; there exists a doubly-stochastic polynomial  $q$  such that  $supp(q) \in supp(p)$  iff  $(1, 1, \dots, 1) \in CO(supp(p))$ .

It follows from Corollary 2.5 that if  $p$  is doubly-stochastic and  $S$ -hyperbolic then  $Prob\{Z_p = E(Z_p)\} > 0$ . And the hyperbolic van der Waerden conjecture can be reformulated as :

*If  $p$  is doubly-stochastic and  $P$ -hyperbolic then*

$$Prob\{|Z_p - E(Z_p)| < \sqrt{2}\} \geq \frac{n!}{n^n}.$$

Perhaps some kind of the measure concentration is present here? ■

**Remark 2.15:** The problem to find out a positive real solution of the inequality (20) is a far reaching generalization of scaling of matrices with nonnegative entries (the corresponding polynomials are multilinear) [20], [23] and scaling of tuples of PSD matrices (the corresponding polynomials are determinantal) [21], [22]. Part 2 of Lemma 2.12 allows to generalize results of [23], [21], [22] to  $P$ -hyperbolic polynomials, even in the black-box setting. Can it be done for all homogeneous polynomials with, say, integer nonnegative coefficients? ■

### 3 The ellipsoid algorithm

Consider a homogeneous polynomial  $q(x), x \in R^n$  of degree  $n$  in  $n$  real variables with nonnegative integer coefficients. Associate with such  $q$  the following convex functional

$$f(y_1, \dots, y_n) = \log(q(e^{y_1}, e^{y_2}, \dots, e^{y_n})).$$

**Proposition 3.1:** *The following conditions are equivalent*

1.  $e = (1, 1, \dots, 1) \in CO(\text{supp}(q))$  .
2.  $\inf_{y_1 + \dots + y_n = 0} f(y_1, \dots, y_n) \geq 0$ .
3. If  $e = (1, 1, \dots, 1) \notin CO(\text{supp}(q))$  then  
 $\inf_{y_1 + \dots + y_n = 0} f(y_1, \dots, y_n) = -\infty$ .  
Let  $\text{dist}(e, CO(\text{supp}(q))) = \Delta^{-1} > 0$  and  $Q = \log(q(e))$  . Define  $\gamma = (Q + 1)\Delta$  . Then

$$\inf_{y_1 + \dots + y_n = 0, (|y_1|^2 + \dots + |y_n|^2)^{\frac{1}{2}} \leq \gamma} f(y_1, \dots, y_n) = \min_{y_1 + \dots + y_n = 0, |y_1|^2 + \dots + |y_n|^2 \leq \gamma} f(y_1, \dots, y_n) \leq -1.$$

**Proof:** Our proof is a straightforward application of the concavity of the logarithm on the positive semi-axis and of the Hanh-Banach separation theorem . It will be included in the full version . ■

Proposition 3.1 suggests the following natural approach to solve Problem 2 , i.e. to decide whether  $e = (1, 1, \dots, 1) \in CO(\text{supp}(q))$  or not :  
find  $\min_{y_1 + \dots + y_n = 0, |y_1|^2 + \dots + |y_n|^2 \leq \gamma} f(y_1, \dots, y_n)$  with absolute accuracy  $\frac{1}{3}$  . If the resulting value is greater than or equal  $-\frac{1}{3}$  then  $e = (1, 1, \dots, 1) \in CO(\text{supp}(q))$  ; if the resulting value is less than or equal  $-\frac{2}{3}$  then  $e = (1, 1, \dots, 1) \notin CO(\text{supp}(q))$  . And , of course , it is natural to use the ellipsoid method . Our main tool is the following property of the ellipsoid algorithm [32]: For a prescribed accuracy  $\delta > 0$ , it finds a  $\delta$ -minimizer of a differentiable convex function  $f$  in a ball  $B$ , that is a point  $x_\delta \in B$  with  $f(x_\delta) \leq \min_B f + \delta$ , in no more than

$$O\left(n^2 \ln\left(\frac{2\delta + \text{Var}_B(f)}{\delta}\right)\right), \quad (\text{Var}_B(f) = \max_B f - \min_B f)$$

iterations. Each iteration requires a single computation of the value and of the gradient of  $f$  at a given point, plus  $O(n^2)$  elementary operations to run the algorithm itself. In our case, this is easily seen to cost at most  $O(n^2)$  oracle calls and  $O(n)$  elementary arithmetic operations .  
We have the  $n - 1$ -dimensional ball  $B_\gamma = \{(y_1, \dots, y_n) : y_1 + \dots + y_n = 0, |y_1|^2 + \dots + |y_n|^2 \leq \gamma\}$ . A straightforward computations show that

$$\text{Var}_B(f) \leq \log(q(1, 1, \dots, 1)e^{\gamma n}) - \log(q(1, 1, \dots, 1)e^{-\gamma n}) \leq 2\gamma n,$$

giving that  $O(n^2(\ln(n) + \ln(\gamma)))$  iterations of the ellipsoid method needed to solve Problem 2 , it amounts to  $O(n^4(\ln(n) + \ln(\gamma)))$  oracle calls . The quantity  $O(n^4(\ln(n) + \ln(\gamma)))$  is polynomial in  $n$  even if  $\gamma$  is exponentially large ( $\text{dist}(e, CO(\text{supp}(q)))$  is exponentially small ). The problem is that if  $\gamma$  is exponentially large ( which can happen ) then we need to call oracles on inputs with exponential bit-size .

Putting things together , we get the following conclusion :

*If it is promised that either  $e = (1, 1, \dots, 1) \in CO(\text{supp}(q))$  or  $\text{dist}(e, CO(\text{supp}(q))) \geq \text{poly}(n)^{-1}$  for some fixed polynomial  $\text{poly}(n)$  then Problem 2 can be solved by a deterministic polynomial-time oracle algorithm based on the ellipsoid method .*

And at this point we can say nothing about Problem 1 , i.e. deciding whether  $e = (1, 1, \dots, 1) \in$

$\text{supp}(q)$  or not . Corollary 2.5 says that if  $q$  is  $S$ -hyperbolic polynomial then Problem 1 and Problem 2 are equivalent ; moreover if  $e = (1, 1, \dots, 1) \notin \text{supp}(q)$  then there exists nonempty  $S \subset \{1, 2, \dots, n\}$  such that

$$\sum_{1 \leq i \leq n} r_i s_i < \sum_{1 \leq i \leq n} s_i = |S| \text{ for all } (r_1, \dots, r_n) \in \text{supp}(q), \quad (21)$$

, where  $(s_1, \dots, s_n)$  is a characteristic function of the subset  $S$  , i.e.  $s_i = 1$  if  $i \in S$  , and  $s_i = 0$  otherwise .

Notice that if (21) holds then the distance  $\text{dist}(e, CO(\text{supp}(q)))$  from the vector  $e = (1, \dots, 1)$  to the Newton polytope  $CO(\text{supp}(q))$  is at least  $\sqrt{\frac{n}{|S|(n-|S|)}} \geq \frac{2}{\sqrt{n}}$  . Thus we have the next theorem .

**Theorem 3.2:** *Problem 1 and Problem 2 are equivalent for  $S$ -hyperbolic polynomials . There exists a deterministic polynomial-time oracle algorithm solving Problem 1 for a given  $S$ -hyperbolic polynomial  $q(\alpha_1, \dots, \alpha_n)$  with integer coefficients . It requires  $O(n^4(\ln(n) + \ln(\ln(q(1, 1, \dots, 1))))$  oracle calls and its bit-wise complexity (which is roughly the radius of the ball  $B_\gamma$  ) is  $O(n^{\frac{1}{2}} \ln(q(1, 1, \dots, 1)))$  .*

## 4 Hyperbolic Sinkhorn scaling

We will discuss briefly in this section another method , which is essentially a large step version of gradient descent .

**Definition 4.1:** Consider an  $e$ -nonnegative tuple  $\mathbf{X} = (x_1, \dots, x_n)$  such that the sum of its components  $S(\mathbf{X}) = d = \sum_{1 \leq i \leq k} x_i$  is  $e$ -positive. Define  $\text{tr}_d(x)$  as a sum of roots of the univariate polynomial equation  $p(x - td) = 0$ .

Define the following map (Hyperbolic Sinkhorn Scaling) acting on such tuples:

$$HS(\mathbf{X}) = \mathbf{Y} = \left( \frac{x_1}{\text{tr}_d(x_1)}, \dots, \frac{x_n}{\text{tr}_d(x_n)} \right)$$

Hyperbolic Sinkhorn Iteration (**HSI**) is the following recursive procedure:

$$\mathbf{X}_{j+1} = HS(\mathbf{X}_j), j \geq 0, \mathbf{X}_0 \text{ is an } e\text{-nonnegative tuple with } \sum_{1 \leq i \leq k} x_i \in C_e .$$

We also define the doubly-stochastic defect of  $e$ -nonnegative tuples with  $e$ -positive sums as

$$DS(\mathbf{X}) = \sum_{1 \leq i \leq k} (\text{tr}_d(x_i) - 1)^2; \sum_{1 \leq i \leq k} x_i = d \in C_e$$

■

We can define the map  $HS(\cdot)$  directly in terms of the  $P$ -hyperbolic polynomial

$$Q(\alpha_1, \dots, \alpha_n) = P_{x_1, \dots, x_n}(\alpha_1, \dots, \alpha_n) = p\left(\sum_{1 \leq i \leq n} \alpha_i x_i\right).$$

Indeed, if  $\sum_{1 \leq i \leq n} \alpha_i x_i = d \in C_e$  then

$$tr_d(\alpha_i x_i) = \frac{\alpha_i \frac{\partial}{\partial \alpha_i} Q(\alpha_1, \dots, \alpha_n)}{Q(\alpha_1, \dots, \alpha_n)} \quad (22)$$

This gives the following way to redefine the map  $HS(\mathbf{X})$  :

$$HS(\alpha_1, \dots, \alpha_n) = \left( \frac{Q(\alpha_1, \dots, \alpha_n)}{\frac{\partial}{\partial \alpha_1} Q(\alpha_1, \dots, \alpha_n)}, \dots, \frac{Q(\alpha_1, \dots, \alpha_n)}{\frac{\partial}{\partial \alpha_n} Q(\alpha_1, \dots, \alpha_n)} \right),$$

for  $\alpha_i > 0, 1 \leq i \leq n$ .

And correspondingly the doubly-stochastic defect of  $(\alpha_1, \dots, \alpha_n)$  is equal to

$$\sum_{1 \leq i \leq n} \left| \frac{\alpha_i \frac{\partial}{\partial \alpha_i} Q(\alpha_1, \dots, \alpha_n)}{Q(\alpha_1, \dots, \alpha_n)} - 1 \right|^2,$$

the same as the left side of (17) . Notice that  $\sum_{1 \leq i \leq n} tr_d(x_i) = n$  by the Euler's identity .

**Example 4.2:** Consider the following hyperbolic polynomial in  $n$  variables:  $p(z_1, \dots, z_n) = \prod_{1 \leq i \leq n} z_i$ . It is  $e$ -hyperbolic for  $e = (1, 1, \dots, 1)$ . And  $N_e$  is a nonnegative orthant,  $C_e$  is a positive orthant. An  $e$ -nonnegative tuple  $\mathbf{X} = (x_1, \dots, x_n)$  can be represented by an  $n \times n$  matrix  $A_{\mathbf{X}}$  with nonnegative entries: the  $i$ th column of  $A$  is a vector  $x_i \in R^n$ . If  $Z = (z_1, \dots, z_n) \in R^n$  and  $d = (d_1, \dots, d_n) \in R^n; z_i > 0, 1 \leq i \leq n$ , then  $tr_d(Z) = \sum_{1 \leq i \leq n} \frac{z_i}{d_i}$ . Recall that for a square matrix  $A = \{a_{ij} : 1 \leq i, j \leq N\}$  row scaling is defined as

$$R(A) = \left\{ \frac{a_{ij}}{\sum_j a_{ij}} \right\},$$

column scaling as  $C(A) = \left\{ \frac{a_{ij}}{\sum_i a_{ij}} \right\}$  assuming that all denominators are nonzero. The iterative process  $\dots CR(CR(A))$  is called *Sinkhorn's iterative scaling* (SI). In terms of the matrix  $A_{\mathbf{X}}$  the map  $HS(\mathbf{X})$  can be realized as follows:

$$A_{HS(\mathbf{X})} = C(R(A_{\mathbf{X}}))$$

So, the map  $HS(\mathbf{X})$  is indeed a (rather far-reaching) generalization of Sinkhorn's scaling. Other generalizations (not all hyperbolic) can be found in [25], [3], [2]. ■

Lemma 2.10 from [19] allows to use (**HSI**) to solve Problem 1 for  $P$ -hyperbolic polynomials  $q$  in the same way as it was done for the perfect matching problem in [25], [20] ; and for the Edmonds' problem in [3] . The corresponding complexity is  $O(n \log(q(e)))$  iterations of (**HSI**) , which can be done in  $O(n^3 \log(q(e)))$  oracle calls . The algorithm works in the following way :  
*Run  $K = O(n \log(q(e)))$  Hyperbolic Sinkhorn Iterations  $\mathbf{X}_{j+1} = HS(\mathbf{X}_j)$  ; if  $DS(\mathbf{X}_i) \leq \frac{1}{n}$  for some  $i \leq K$  then the  $p$ -mixed form  $M_p(\mathbf{X}_0) > 0$  , and  $M_p(\mathbf{X}_0) = 0$  otherwise .*

## 5 Half-Plane Property

The following definition is from [7].

**Definition 5.1:** A polynomial  $P(z_1, \dots, z_n)$  in  $n$  complex variables is said to have the "half-plane property" if  $P(z_1, \dots, z_n) \neq 0$  provided  $\operatorname{Re}(z_i) > 0$ . ■

In a control theory literature (see [36]) the same property is called *Wide sense stability*. And *Strict sense stability* means that

$P(z_1, \dots, z_n) \neq 0$  provided  $\operatorname{Re}(z_i) \geq 0$ .

The following simple fact shows that for homogeneous polynomials the "half-plane property" is, up to a single factor, the same as  $P$ -hyperbolicity.

**Fact 5.2:** A homogeneous polynomial  $R(z_1, \dots, z_n)$  has the "half-plane" property if and only if there exists real  $\alpha$  such that the polynomial  $e^{i\alpha}R(z_1, \dots, z_n)$  is  $P$ -hyperbolic polynomial with real nonnegative coefficients. ■

**Proof:**

1. Suppose that  $R(z_1, \dots, z_n) = e^{-i\alpha}Q(z_1, \dots, z_n)$  where  $\alpha$  is real and  $Q$  is  $P$ -hyperbolic. Then  $Q$  is  $(1, 1, \dots)$ -hyperbolic and all real vectors  $(x_1, \dots, x_n)$  with positive coordinates are  $(1, 1, \dots)$ -positive. Therefore  $Q$  is  $(x_1, \dots, x_n)$ -hyperbolic for all real vectors  $(x_1, \dots, x_n) \in R_{++}^n$  with positive coordinates. It follows that  $|R(x_1 + iy_1, \dots, x_n + iy_n)| = |Q(x_1 + iy_1, \dots, x_n + iy_n)| = |Q(x_1, \dots, x_n) \prod_{1 \leq k \leq n} (1 + i\lambda_k)|$ , where  $(\lambda_1, \dots, \lambda_n)$  are real roots of the real vector  $(y_1, \dots, y_n)$  in the direction  $(x_1, \dots, x_n)$ .

This gives the following inequality, which is equivalent to the "half-plane property" of  $R$ :

$$\begin{aligned} |R(x_1 + iy_1, \dots, x_n + iy_n)| &\geq |R(x_1, \dots, x_n)| = \\ &= |Q(x_1, \dots, x_n)| > 0 : \\ (x_1, \dots, x_n) &\in R_{++}^n, (y_1, \dots, y_n) \in R^n \end{aligned} \tag{23}$$

2. Suppose that  $R(z_1, \dots, z_n)$  has the "half-plane property" and consider the roots of the following polynomial equation in one complex variable:  $P(x_1 - z, x_2 - z, \dots, x_n - z) = 0$ , where  $(x_1, \dots, x_n) \in R^n$  is a real vector,  $z = x + iy \in C$ . If the imaginary part  $\operatorname{Im}(z) = y$  is not zero then, using the homogeneity,  $R(i\frac{x-x_1}{y} + 1, \dots, i\frac{x-x_n}{y} + 1) = 0$ , which is impossible as  $R$  has the "half-plane property". Therefore all roots of  $R(X - te) = 0$  are real for all real vectors  $X \in R^n$  (here  $e = (1, 1, \dots, 1)$ ). In the same way all roots of  $R(X - te) = 0$  are real positive numbers if  $X \in R_{++}^n$ . It follows that if  $X \in R^n$  then  $R(X) = R(e) \prod_{1 \leq k \leq n} \lambda_k(X)$ , where  $(\lambda_1, \dots, \lambda_n)$  are (real) roots of the equation  $R(X - te) = 0$ . Thus the polynomial  $(\frac{1}{R(e)})R$  takes real values on  $R^n$  and therefore its coefficients are real. In other words, the polynomial  $(\frac{1}{R(e)})R$  is  $P$ -hyperbolic. If  $R(1, 1, \dots, 1) = e^{-i\alpha}|R(1, 1, \dots, 1)|$  then the polynomial  $e^{i\alpha}R$  is also  $P$ -hyperbolic. (Recall that the coefficients of any  $P$ -hyperbolic polynomial  $p$  are nonnegative for they are  $p$ -mixed forms of  $e$ -nonnegative tuples, and  $p$ -mixed forms of  $e$ -nonnegative tuples are nonnegative if  $p(e) > 0$  [26].)



■

We use this observation to show that Theorem 2.2 in this paper implies (and seriously strengthens) Theorem 7.2 in [7] , which is the main result of a very long recent paper [7] .

## 5.1 Submodularity and hyperbolicity

Let  $p$  be a  $P$ -hyperbolic polynomial of degree  $n$  in  $n$  variables . It follows from Theorem 2.2 that  $r = (r_1, r_2, \dots, r_n) \in \text{supp}(p)$  if and only if the following inequalities hold :

$$r(S) = \sum_{i \in S} r_i \leq R(S) = \text{Rank}_p\left(\sum_{i \in S} e_i\right); S \subset \{1, 2, \dots, n\}.$$

**Fact 5.3:** The functional  $R(S) = \text{Rank}_p(\sum_{i \in S} e_i)$  is normalized , i.e.  $R(\emptyset) = 0$  , and submodular , i.e.  $R(A \cup B) \leq R(A) + R(B) - R(A \cap B)$  :  $A, B \subset \{1, 2, \dots, n\}$  . ■

**Proof:** Associate with two subsets  $A, B \subset \{1, 2, \dots, n\}$  the following three  $e$ -nonnegative vectors :

$$x = \sum_{i \in A \setminus (A \cap B)} e_i, y = \sum_{i \in A \cap B} e_i, z = \sum_{i \in B \setminus (A \cap B)} e_i.$$

We need to prove the inequality  $\text{Rank}_p(x + y + z) \leq \text{Rank}_p(x) + \text{Rank}_p(z) - \text{Rank}_p(y)$ . This inequality is obvious and well known for positive semidefinite matrices . The extension to  $e$ -nonnegative vectors respect to  $e$ -hyperbolic polynomial  $p$  is done in the same way as in the proof of Corollary A.3 : consider a hyperbolic in the direction  $(1, 1, 1)$  polynomial

$$L(\alpha_1, \alpha_2, \alpha_3) = M_p(k, \dots, k, e, \dots, e), k = \alpha_1 x + \alpha_2 y + \alpha_3 z;$$

where the vectors  $x, y, z$  are  $e$ -nonnegative respect to hyperbolic polynomial  $p$  , and the tuple  $(k, \dots, k, e, \dots, e)$  consists of  $\text{Rank}_p(x + y + z)$  copies of  $k$  and  $n - \text{Rank}_p(x + y + z)$  copies of  $e$ . After that apply Theorem 1.5 . ■

### Corollary 5.4:

1. A support  $\text{supp}(p)$  of  $P$ -hyperbolic polynomial  $p$  is an intersection of the integral polymatroid  $\{(r_1, \dots, r_n) : r(S) = \sum_{i \in S} r_i \leq R(S) = \text{Rank}_p(\sum_{i \in S} e_i); S \subset \{1, 2, \dots, n\}\}$  with the hyperplane  $\{(r_1, \dots, r_n) : \sum_{1 \leq i \leq n} r_i = n\}$ .
2. A support  $\text{supp}(R)$  of any polynomial  $R$  with the "half-plane" property is a jump system .

**Proof:** (Consult [24] for a definition and some properties of jump systems and integral polymatroids ) . This Corollary follows directly Theorem 2.2 , Fact 5.3 and Proposition (3.1) in [24] . ■

It is quite amazing how the two communities , "hyperbolic" and "half-plane" , were not aware about each other results for a long , long time . (Interestingly , two authors of [7] and one author of [27] were with the same department until very recently . Perhaps , one needs to be a dilettante to notice a bridge .)

## 6 Conclusion and Acknowledgments

Univariate polynomials with real roots appear quite often in modern combinatorics , especially in the context of integer polytopes . We discovered in this paper rather unexpected and very likely far-reaching connections between hyperbolic polynomials and many classical combinatorial and algorithmic problems . (The author taught about "On hyperbolic nature of perfect marriages" as a title of this paper , but with the current climate it could be understood in many ways .) There are still several open problems . The most interesting is Conjecture 2.11 in this paper , which is a generalization of the van der Waerden conjecture for permanents of doubly stochastic matrices and many others related questions .

For a hyperbolic in direction  $(1, 1, \dots, 1)$  polynomial  $Mul(y_1, \dots, y_n) = y_1 y_2 \dots y_n$  Conjecture 2.11 is equivalent to the famous van der Waerden conjecture for permanents of doubly stochastic matrices , proved in [15] , [16] . For a hyperbolic in direction  $I$  polynomial  $\det(X)$  ,  $X$  is  $n \times n$  hermitian matrix , it is equivalent to Bapat's conjecture [5] (it was also hinted in [15] ) , proved by the author in [21] , [35] . It also holds for the Moore determinant  $Det(M)(Y)$  ,  $Y$  is  $n \times n$  quaternionic hermitian matrix , with the proof essentially the same as in [35] .

Another , equivalent form of "hyperbolic" (or "half-plane" ) van der Waerden conjecture can be formulated as follows :

**Conjecture 6.1:** Consider a homogeneous polynomial  $p(z_1, \dots, z_n)$  of degree  $n$  in  $n$  complex variables . Assume that this polynomial satisfies the property :

$$|p(z_1, \dots, z_n)| \geq \prod_{1 \leq i \leq n} Re(z_i) \text{ on the domain } \{(z_1, \dots, z_n) : Re(z_i) \geq 0, 1 \leq i \leq n\} .$$

Is it true that  $|\frac{\partial^n}{\partial z_1 \dots \partial z_n} p| \geq \frac{n!}{n^n}$  ? .

(Notice that Theorem 2.10 and Fact 5.2 imply that  $\frac{\partial^n}{\partial z_1 \dots \partial z_n} p \neq 0$  .) ■

It would be very interesting and enlighting to prove Conjecture 2.11 using methods of the theory of functions of many complex variables. Fact 5.2 , together with other results in this paper , makes a connection between the Complexity Theory and the theory of linear time-multidimensional systems : all "hard" instances of Problem 1 are necessary unstable polynomials.

Another interesting conjecture is related to the majorization :

**Conjecture 6.2:** Consider the doubly-stochastic and  $P$ -hyperbolic homogeneous polynomial  $p(x_1, \dots, x_n)$  of degree  $n$  in  $n$  real variables .

Let  $\Lambda(X) \in R^n$  be a real  $n$ -dimensional vector , whose coordinates are the roots of the equation  $p(X - te) = 0$  , where  $X \in R^n$  and  $e$  is the vector of all ones . Then there exists a  $n \times n$  doubly stochastic matrix  $A$  such that  $\Lambda(X) = AX$ .

(Some partial and related results in this direction can be found in [19] ; this conjecture is true for determinantal polynomials .) ■

A natural extension of **Problem 1** is for  $P$ -hyperbolic polynomials to approximate  $\frac{\partial^n}{\partial x_1 \dots \partial x_n} p(x_1, \dots, x_n)$  within a multiplicative factor using deterministic ( or randomized ) polynomial-time oracle algorithms . It is not clear to the author whether known recent randomized algorithms for  $(1 + \epsilon)$  approximation of the permanent  $Per(B)$  of entry-wise nonnegative matrix  $B$  can be done in the "oracle fashion" , i.e. using only some outputs of the multilinear polynomial

$$q(x_1, \dots, x_n) = \prod_{1 \leq i \leq n} \sum_{1 \leq j \leq n} B(i, j) x_j.$$

If hyperbolic van der Waerden conjecture is true then the technique in this paper , similarly to [20] and [21] , [22] , would produce a deterministic polynomial-time oracle algorithm with  $\frac{n^n}{n!}$  multiplicative factor .

The technique developed in this paper can be applied to other "noble" desicion problems . For instance , checking factorizability of  $P$ -hyperbolic polynomials can be also done in deterministic oracle polynomial time . The factorizability is closely related to the hyperbolic generalization of the indecomposability of matrix tuples [22].

This paper is probably the first one which uses Theorem 1.5 in the combinatorial context . We expect many more such applications of Theorem 1.5 . This (very nontrivial) theorem , when in good hands , is a powerful tool allowing reasonably simple and short proofs .

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## A Proof of the (main ) Theorem 2.2

Before proving Theorem 2.2 , we will recall some basic properties of  $p$ -mixed forms and prove a few auxillary results . The following fact was proved in [26]

**Fact A.1:** Consider a homogeneous polynomial  $p(x), x \in R^m$  of degree  $n$  in  $m$  real variables which is hyperbolic in the direction  $e$ . Then the following properties hold .

1. The  $p$ -mixed form  $M_p(x_1, \dots, x_n)$  is linear in each  $x_i, 1 \leq i \leq n$ .
2. If  $x_1, x_2, \dots, x_{n-1}$  are  $e$ -nonnegative then the linear functional  $l(x) = M_p(x_1, \dots, x_{n-1}, x)$  is nonnegative on the closed cone  $N_e$  of  $e$ -nonnegative vectors .
3. If the tuples  $(x_1, \dots, x_n), (y_1, \dots, y_n), (x_1 - y_1, \dots, x_n - y_n)$  are  $e$ -nonnegative then

$$0 \leq M_p(y_1, \dots, y_n) \leq M_p(x_1, \dots, x_n).$$

4. Fix  $e$ -positive vector  $d$  and consider the following homogeneous polynomial  $p_d(x), x \in R^m$  of degree  $n - 1$  in  $m$  real variables :  $p_d(x) =: M_p(x, x, \dots, x, d)$  . Then  $p_d$  is hyperbolic in any  $e$ -positive direction  $v \in C_e(p)$  . If  $g \in C_e(p)$  (  $e$ -positive respect to the polynomial  $p$  ) then also  $q \in C_v(p_d)$  for all  $v \in C_e(p)$  .

■

The next fact is well known .

**Fact A.2:** Consider a sequence of univariate polynomials of the same degree  $n$  :  $P_k(t) = \sum_{0 \leq i \leq n} a_{i,k} t^i$  . suppose that  $\lim_{k \rightarrow \infty} a_{i,k} = a_i, 0 \leq i \leq n$  and  $a_n \neq 0$  . Define  $P(t) = \sum_{0 \leq i \leq n} a_i t^i$  . Then roots of  $P_k$  converge to roots of  $P$  . In particular if roots of all polynomials  $P_k$  are real then also roots of  $P$  are real ; if roots of all polynomials  $P_k$  are real nonnegative numbers then also roots of  $P$  are real nonnegative numbers . ■

The following corollary of Theorem 1.5 plays crucial role in our proof of Theorem 2.2 .

**Corollary A.3:**

1. Consider a homogeneous polynomial  $p(x), x \in R^m$  of degree  $n$  in  $m$  real variables which is hyperbolic in the direction  $e$ . Let  $x_1, x_2, x_3$  be three  $e$ -nonnegative vectors and  $d = x_1 + x_2 + x_3$  is  $e$ -positive . Assume wlog that  $p(x_1 + x_2 + x_3) = 1$  . Then there exists three symmetric positive semidefinite matrices  $A, B, C$  such that  $p(a_1 x_1 + a_2 x_2 + a_3 x_3) = \det(a_1 A + a_2 B + a_3 C)$  for all real  $a_1, a_2, a_3$ . Additionally , the roots of  $a_1 x_1 + a_2 x_2 + a_3 x_3$  in the direction  $d$  , i.e. the roots of the equation  $p(a_1 x_1 + a_2 x_2 + a_3 x_3 - t d) = 0$  , coincide with the eigenvalues of  $a_1 A + a_2 B + a_3 C$  .
2. Theorem 2.2 is true for  $e$ -nonnegative tuples  
 $(\mathbf{X}) = (x_1, \dots, x_n), x_i \in R^m$  consisting of at most three distinct components , i.e the cardinality of the set  $\{x_1, \dots, x_n\}$  is at most three .

**Proof:**

1. Consider the following homogeneous polynomial  $L(b_1, b_2, b_3) = p(b_1x_1 + b_2x_2 + b_3(x_1 + x_2 + x_3))$  of degree  $n$  in 3 real variables . It follows from Theorem 1.5 that there exists two real symmetric matrices  $A$  and  $B$  such that  $L(b_1, b_2, b_3) = \det(b_1A + b_2B + b_3I)$  . It follows that they both positive semidefinite , and  $C = I - A - B$  is also positive semidefinite . Take a real linear combination  $z = a_1x_1 + a_2x_2 + a_3x_3$ . Then

$$\begin{aligned} p(z - t(x_1 + x_2 + x_3)) &= \\ \det((a_1 - a_3)A + (a_2 - a_3)B + a_3I - tI) &= \\ = \det(a_1A + a_2B + a_3C - tI). \end{aligned}$$

This proves that  $p(a_1x_1 + a_2x_2 + a_3x_3) = \det(a_1A + a_2B + a_3C)$  for all real  $a_1, a_2, a_3$  by putting  $t = 0$ . And it also proves the "eigenvalues" statement .

2. Consider  $e$ -nonnegative tuple  $(\mathbf{X})$  consisting of  $r_i$  copies of  $x_i$  ,  $1 \leq i \leq 3$  ;  $r_1 + r_2 + r_3 = n$  . Assume that  $d = x_1 + x_2 + x_3$  is  $e$ -positive (if it is not then  $M_p(\mathbf{X}) = 0$  by a simple argument based on the monotonicity of  $p$ -mixed forms ). It follows from the polarization formula (10) that

$$M_p(\mathbf{X}) = \sum_{1 \leq i \leq k < \infty} d_i p(t_{1,i}x_1 + t_{2,i}x_2 + t_{3,i}x_3),$$

and this formula is universal , i.e. holds for all homogeneous polynomial of degree  $n$  , in particular for  $\det(X)$  ,  $X$  is  $n \times n$  symmetric matrix . Therefore , using the first part of this Corollary we get that the  $p$ -mixed form  $M_p(\mathbf{X}) = D(\mathbf{A})$  , where the matrix tuple  $\mathbf{A}$  consists of  $r_1$  copies of  $A$  ,  $r_2$  copies of  $B$  and  $r_3$  copies of  $C$  and  $D(\mathbf{A})$  is the mixed discriminant . Using Rado theorem for mixed discriminants we get that  $D(\mathbf{A}) > 0$  iff

$$\text{Rank}(\sum_{i \in S} A_i) \geq \sum_{i \in S} r_i \text{ for all } S \subset \{1, 2, 3\}.$$

But from the first part we get that  $\text{Rank}(\sum_{i \in S} A_i)$  is equal to  $p$ -rank  $\text{Rank}_p(\sum_{i \in S} x_i)$  of  $\sum_{i \in S} x_i$  for all  $S \subset \{1, 2, 3\}$  .

■

**Proposition A.4:** Consider similarly to part 4 of Fact A.1 the polynomial  $p_d(x) =: M_p(x, x, \dots, x, d)$  where  $d$  is  $e$ -nonnegative and  $\text{Rank}_p(d) \geq 1$  . Then  $p_d$  is hyperbolic in any direction  $z \in N_e(p)$  which is  $e$ -nonnegative and satisfies the following inequalities :

$$\text{Rank}_p(z) \geq n - 1; \text{Rank}_p(z + d) = n. \quad (24)$$

Also , if  $y \in N_e(p)$  is  $e$ -nonnegative then also  $y \in N_z(p_d)$  , i.e. is  $z$ -nonnegative respect to the polynomial  $p_d$ .

**Proof:** Let  $z \in N_e(p)$  be  $e$ -nonnegative vector satisfying (24) . Consider univariate polynomial  $P(t) = M_p(tz + x, tz + x, \dots, tz + x, d)$  . Then  $P(t) = \sum_{0 \leq i \leq n-1} a_i t^i$  and  $a_{n-1} = M_p(z, z, \dots, z, d)$  . It follows from Corollary A.3 that  $a_{n-1} > 0$  . Consider now a sequence of univariate polynomials

$P_k(t) = M_p(tz_k + x, tz_k + x, \dots, tz_k + x, d_k)$  . Where  $z_k, d_k$  are  $e$ -positive and  $\lim_{k \rightarrow \infty} z_k = z$ ,  $\lim_{k \rightarrow \infty} d_k = d$  . Then the coefficients of polynomials  $P_k$  converge to the coefficients of the polynomial  $P$  . It follows from part 4 of Fact A.1 that the roots of  $P_k$  are real . Since  $a_{n-1} > 0$  hence using Fact A.2 we get that the roots of  $P$  are also real . This exactly means that the polynomial  $p_d$  is hyperbolic in direction  $z$  . The  $d$ -nonnegativity statement follows from the nonnegativity part of Fact A.2 . ■

We are ready now to present our proof of Theorem 2.2 . The proof is by induction in the degree  $n$  . The main trick which we used is that to justify the induction , i.e. that if the generalized Rado conditions hold for hyperbolic polynomial  $p$  of degree  $n$  then the generalized Rado conditions hold for some auxillary hyperbolic polynomial  $p_d$  of degree  $n - 1$  , we need to prove Theorem 2.2 for tuples consisting of at most three distinct components . And this particular case follows from the classical Rado theorem via Theorem 1.5 and Corollary A.3 .

**Proof: (Proof of Theorem 2.2) .**

The "only if" part is simple . Indeed supposed that there exists a subset  $S \subset \{1, 2, \dots, n\}$  such that  $\text{Rank}_p(\sum_{i \in S} x_i) < |S|$  , i.e. using the identities (14)  $M_p(k, k, \dots, k, d, \dots, d) = 0$  , where  $k = \sum_{i \in S} x_i$  ,  $d \in C_e(p)$  is  $e$ -positive and the  $n$ -tuple  $(k, k, \dots, k, d, \dots, d)$  consists of  $|S|$  copies of  $k = \sum_{i \in S} x_i$  . Let  $d$  be any  $e$ -positive positive vector such that  $d - x_i$  is  $e$ -nonnegative ,  $1 \leq i \leq n$  . Using the monotonicity of  $p$ -mixed forms we get that

$$M_p(x_1, \dots, x_n) \leq M_p(k, k, \dots, k, d, \dots, d) = 0.$$

Our proof of the "if" part is by induction in the degree  $n$  . Suppose that the generalized Rado conditions (15) hold . Then at least  $\text{Rank}_p(x_n) \geq 1$  . Consider the following homogeneous polynomial of degree  $n - 1$  :

$$p_d(x) = M_p(x, x, \dots, x, d), \quad d = x_n.$$

We get from Proposition A.4 the following assertion :

The polynomial  $p_d(x)$  is hyperbolic in direction  $z = \sum_{1 \leq i \leq n-1} x_i$  and the vectors  $x_i \in N_z(p_d)$ ,  $1 \leq i \leq n - 1$  , i.e. are  $z$ -nonnegative respect to the polynomial  $p_d$ .

Indeed , it follows from the generalized Rado conditions (15) that  $\text{Rank}_p(z) \geq n - 1$  and  $\text{Rank}_p(z + d) = \text{Rank}_p(\sum_{1 \leq i \leq n} x_i) = n$  .

Next we show that the  $n-1$ -tuple  $\mathbf{Y} = (x_1, \dots, x_{n-1})$  satisfies the generalized Rado conditions for  $z$ -hyperbolic polynomial  $p_d$  of degree  $n - 1$  :

$$\text{Rank}_{p_d}(\sum_{i \in S} x_i) \geq |S| \quad \text{for all } S \subset \{1, 2, \dots, n - 1\}.$$

Or equivalently (see formulas (14) ) , that

$$M_p(k, \dots, k, z, \dots, z, d) > 0; k = \sum_{i \in S} x_i, \quad (25)$$

$$z = \sum_{1 \leq i \leq n-1} x_i, d = x_n, S \subset \{1, \dots, n - 1\}, \quad (26)$$

where the  $n$ -tuple  $\mathbf{T} = (k, \dots, k, z, \dots, z, d)$  consists of  $|S|$  copies of  $k$  ,  $n - 1 - |S|$  copies of  $z$  and one copy of  $d$  .



It is easy to see that the generalized Rado conditions for the  $n$ -tuple  $\mathbf{T}$  are implied by the generalized Rado conditions for the original  $n$ -tuple  $\mathbf{X} = (x_1, \dots, x_{n-1}, x_n)$ . Since the  $n$ -tuple  $(k, \dots, k, z, \dots, z, d)$  consists of at most three distinct components hence we can apply part 2 of Corollary A.3. Therefore we get that indeed

$M_p(k, \dots, k, z, \dots, z, d) > 0$  and hence the following inequalities hold :

$$\text{Rank}_{p_d}(\sum_{i \in S} x_i) \geq |S| \text{ for all } S \subset \{1, 2, \dots, n-1\}. \quad (27)$$

Thus, by induction in the degree, we get that  $p_d$ -mixed form  $M_{p_d}(x_1, \dots, x_{n-1}) > 0$  : the polynomial  $p_d$  of degree  $n-1$  in  $m$  real variables is  $z$ -hyperbolic. But

$$\begin{aligned} M_{p_d}(x_1, \dots, x_{n-1}) &= \frac{\partial^{n-1}}{\partial \alpha_1 \dots \partial \alpha_{n-1}} p_d(\sum_{1 \leq i \leq n-1} \alpha_i x_i) = \\ &= \frac{\partial^{n-1}}{\partial \alpha_1 \dots \partial \alpha_{n-1}} M_p(\sum_{1 \leq i \leq n-1} \alpha_i x_i, \dots, \sum_{1 \leq i \leq n-1} \alpha_i x_i, x_n) = (n-1)! M_p(x_1, \dots, x_n). \end{aligned}$$

We conclude that if Theorem 2.2 is true for  $n-1$  then it is also true for  $n$ , and the case " $n=1$ " is trivially true. ■

**Remark A.5:** Consider a mixed discriminant  $D(\mathbf{A})$ , where  $\mathbf{A} = (A_1, \dots, A_n)$  is a  $n$ -tuple of positive semidefinite  $n \times n$  hermitian matrices, i.e.  $A_i \succeq 0$ . Recall that in this case  $D(\mathbf{A}) \geq 0$ ; and  $D(\mathbf{A}) > 0$  if and only if there exists  $n$  linearly independent vectors  $v_1, \dots, v_n$  such that  $v_i \in \text{Im}(A_i)$ ,  $1 \leq i \leq n$ .

In the proof of Theorem 2.2 we encountered the following tuple of positive semidefinite matrices :

$\mathbf{A} = (A, \dots, A, B, \dots, B, C)$  consisting of  $l$  copies of  $A$ ,  $m$  copies of  $B$  and one copy of  $C$ . Moreover, this tuple is even more special. I.e.  $B - A \succeq 0$ ,  $\text{Rank}(B) \geq n-1$ ,  $\text{Rank}(A) \geq l$ ,  $\text{rank}(C) \geq 1$ ,  $\text{Rank}(A+C) \geq l+1$  and  $\text{Rank}(B+C) = n$ .

For such tuples the Rado theorem has very elementary proof, which we sketch below.

There are two cases. First case is when  $\text{Rank}(B) = n$ , it is simple and left to the reader. Second case is when  $\text{Rank}(B) = n-1$ .

This is how in this case we can choose vectors  $v_n \in \text{Im}(C)$ ;  $v_1, \dots, v_l \in \text{Im}(A)$ ;  $v_{l+1}, \dots, v_{n-1} \in \text{Im}(B)$  in such a way that  $(v_1, \dots, v_n)$  is a basis : first choose nonzero  $v_n \in \text{Im}(C)$  which does not belong to  $\text{Im}(B)$ , second choose any  $l$  linearly independent vectors  $v_1, \dots, v_l \in \text{Im}(A)$ , third choose any  $n-l-1$  linearly independent vectors in  $\text{Im}(B) \cap L(v_1, \dots, v_l)^\perp$ . ( $L(v_1, \dots, v_l)^\perp$  is a linear subspace orthogonal to the linear subspace  $L(v_1, \dots, v_l)$  which is spanned by  $(v_1, \dots, v_l)$ .)

■

## B Proof of Proposition 2.6

**Proof:** Assume wlog that  $q(\alpha_1, \dots, \alpha_n) = 1$ . It follows from the Euler's identity that

$$\sum_{1 \leq i \leq n} \alpha_i \frac{\partial}{\partial \alpha_i} q(\alpha_1, \dots, \alpha_n) = n.$$

Let  $q(\alpha_1, \dots, \alpha_n) = \sum_{(r_1, \dots, r_n) \in \text{supp}(q)} a_{(r_1, \dots, r_n)} \prod_{1 \leq i \leq n} \alpha_i^{r_i}$ .  
Define the following nonnegative real numbers :

$$b_{(r_1, \dots, r_n)} = a_{(r_1, \dots, r_n)} \prod_{1 \leq i \leq n} \alpha_i^{r_i}, (r_1, \dots, r_n) \in \text{supp}(q).$$

Then  $\alpha_i \frac{\partial}{\partial \alpha_i} q(\alpha_1, \dots, \alpha_n) = \sum_{(r_1, \dots, r_n) \in \text{supp}(q)} r_i b_{(r_1, \dots, r_n)}$ .

Suppose that for some subset  $S \subset \{1, 2, \dots, n\}$ ,  $1 \leq |S| < n$  we have the inequality  $\sum_{i \in S} r_i < |S|$  for all  $(r_1, \dots, r_n) \in \text{supp}(q)$ . Then  $\sum_{i \in S} \alpha_i \frac{\partial}{\partial \alpha_i} q(\alpha_1, \dots, \alpha_n) \leq |S| - 1$ . But the condition (17) says that  $\alpha_i \frac{\partial}{\partial \alpha_i} q(\alpha_1, \dots, \alpha_n) = 1 + \delta_i$  and  $\sum_{1 \leq i \leq n} |\delta_i|^2 \leq \frac{1}{n}$ . By the Cauchy-Schwarz inequality,  $\sum_{i \in S} |\delta_i| \leq \sqrt{\frac{|S|}{n}} < 1$ . Therefore,

$$\sum_{i \in S} \alpha_i \frac{\partial}{\partial \alpha_i} q(\alpha_1, \dots, \alpha_n) \geq |S| - \sum_{i \in S} |\delta_i| > |S| - 1.$$

The last inequality gives a contradiction. ■

## C A sketch of a proof of Corollary 2.5

**Proof:** By Theorem 2.2 the conditions (1) and (2) are equivalent. (2) implies (3) for any homogeneous polynomial with nonnegative coefficients.

Let  $\alpha_i = e^{y_i}$ ,  $1 \leq i \leq n$ ;  $\sum_{1 \leq i \leq n} y_i = 0$ . Consider the following convex functional

$$f(y_1, \dots, y_n) = \log(q(e^{y_1}, e^{y_2}, \dots, e^{y_n})).$$

Here  $q(x)$ ,  $x \in \mathbb{R}^n$  is a homogeneous polynomial of degree  $n$  in  $n$  real variables with nonnegative coefficients. Then

$$\frac{\alpha_i \frac{\partial}{\partial \alpha_i} q(\alpha_1, \dots, \alpha_n)}{q(\alpha_1, \dots, \alpha_n)} = \frac{\partial}{\partial y_i} f(y_1, \dots, y_n), 1 \leq i \leq n.$$

Notice the condition (3) is equivalent to the following condition :

$$\inf_{y_1 + \dots + y_n = 0} f(y_1, \dots, y_n) = L > -\infty.$$

Consider the anti-gradient flow, i.e. the system of differential equations

$$y_i(t)' = -\left(\frac{\partial}{\partial y_i} f(y_1, \dots, y_n) - 1\right), y_i(0) = 0; 1 \leq i \leq n.$$

It is well known that in this convex case the gradient flow is defined for all  $t \geq 0$ . Using the Euler's identity we get that

$$\frac{d}{dt} f(y_1(t), \dots, y_n(t)) = -\beta(t) =: -\sum_{1 \leq i \leq n} \left| \frac{\alpha_i \frac{\partial}{\partial \alpha_i} q(\alpha_1, \dots, \alpha_n)}{q(\alpha_1, \dots, \alpha_n)} - 1 \right|^2$$

It is easy to see that , because of the convexity of  $f$  , a nonnegative function  $\beta(t)$  is non-increasing on  $[0, \infty)$  .

As  $\inf_{y_1+\dots+y_n=0} f(y_1, \dots, y_n) = L > -\infty$  thus  $\int_0^\infty \beta(t) dt < \infty$  . Thus  $\lim_{t \rightarrow \infty} \beta(t) = 0$  . This proves the implication (3)  $\rightarrow$  (4) for all homogeneous polynomials of degree  $n$  in  $n$  real variables with nonnegative coefficients .

The implication (4)  $\rightarrow$  (5) is obvious . The implication (5)  $\rightarrow$  (6) for general homogeneous polynomials of degree  $n$  in  $n$  real variables with nonnegative coefficients is Proposition 2.6 .

Finally , the implication (6)  $\rightarrow$  (2) follows fairly directly from Theorem 2.2 . ■

## D Lower bounds on the number of oracle calls for the exact computation of $\frac{\partial^n}{\partial x_1 \dots \partial x_n} p(x_1, \dots, x_n)$

**Definition D.1:** Call a set  $\{X_1, \dots, X_m\}, X_i \in C^n$   $\epsilon$ -universal if there exist complex numbers  $c_1, \dots, c_m$  such that for any homogeneous polynomial  $p(\cdot)$  of degree  $n$  in  $n$  complex variables the following inequality holds

$$\begin{aligned} & \left| \frac{\partial^n}{\partial x_1 \dots \partial x_n} p(x_1, \dots, x_n) - \sum_{1 \leq i \leq m} c_i p(X_i) \right| \\ & \leq \epsilon \max_{(r_1, \dots, r_n) \in I_{n,n}} |a_{r_1, \dots, r_n}|, \end{aligned} \quad (28)$$

where  $a_{r_1, \dots, r_n}, (r_1, \dots, r_n) \in I_{n,n}$  are the coefficients of the polynomial  $p(\cdot)$  . ■

**Lemma D.2:** If the set  $\{X_1, \dots, X_m\}, X_i \in C^n$  is 0-universal then

$$m \geq \frac{n!}{[\frac{n}{2}]!(n - [\frac{n}{2}])!} \approx \frac{2^n}{\sqrt{n}} \quad (29)$$

If the set  $\{X_1, \dots, X_m\}, X_i \in C^n$  is  $\epsilon$ -universal then

$$m \geq \min\left(\left\lceil \frac{1}{\epsilon} \right\rceil, \frac{n!}{[\frac{n}{2}]!(n - [\frac{n}{2}])!}\right) \quad (30)$$

**Proof:** Define a monomial  $M_{r_1, \dots, r_n}(x_1, \dots, x_n) = x_1^{r_1} x_2^{r_2} \dots x_n^{r_n}$  . As  $\{X_1, \dots, X_m\}$  is universal thus there exists complex numbers  $(c_1, \dots, c_m)$  , which are wlog are all nonzero , such that

$$\sum_{1 \leq i \leq m} c_i M_{r_1, \dots, r_n}(c_i^{\frac{1}{n}} X_i) = 0$$

if  $(r_1, \dots, r_n) \in I(n, n), (r_1, \dots, r_n) \neq (1, 1, \dots, 1)$  ;

and  $\sum_{1 \leq i \leq m} M_{1, 1, \dots, 1}(c_i^{\frac{1}{n}} X_i) = 1$  ; define  $Y_i = c_i^{\frac{1}{n}} X_i$  (here  $c_i^{\frac{1}{n}}$  is one of the  $n$ th complex roots of  $c_i$  ).

Let  $Half = \{(r_1, \dots, r_n) : r_i \in \{0, 1\}, \sum_{1 \leq i \leq n} r_i = [\frac{n}{2}]\}$  . Notice that the cardinality  $|Half| = \frac{n!}{[\frac{n}{2}]!(n - [\frac{n}{2}])!} =: K$  .

Define the following two  $K \times m$  complex matrices :

$$W((r_1, \dots, r_n), j) = M_{r_1, \dots, r_n}(Y_j),$$

$$V((r_1, \dots, r_n), j) = M_{1-r_1, \dots, 1-r_n}(Y_j) : \\ (r_1, \dots, r_n) \in Half, 1 \leq j \leq m.$$

Clearly ,  $Rank(W) = Rank(V) \leq m$  . On the other hand the 0-universality condition implies the matrix identity

$$WV^T = I \quad (31)$$

Therefore  $m \geq Rank(W) \geq |Half| = \frac{n!}{[\frac{n}{2}]!(n-[\frac{n}{2}])!}$  .

If the set  $\{X_1, \dots, X_m\}$  is  $\epsilon$ -universal and  $d\epsilon < 1, d \in N$  then  $Rank(WV^T) \geq d$  . This proves (30) . ■

**Remark D.3:** The identity (7) is a particular case of a slightly more general one :

$$\frac{\partial^n}{\partial x_1 \dots \partial x_N} p(x_1, \dots, x_n) = E(p(z_1, z_2, \dots, z_n) \prod_{1 \leq i \leq n} \overline{z_i}), \quad (32)$$

where  $(z_1, z_2, \dots, z_n)$  are independent complex random variables such that  $E(z_i) = 0$  and  $E(z_i \overline{z_i}) = 1$  for all  $1 \leq i \leq n$  . The identity is easily proved by checking it for all monomials  $M_{r_1, \dots, r_n}, (r_1, \dots, r_n) \in I(n, n)$  . If  $p(\cdot)$  is a multilinear polynomial , i.e.

$$p(x_1, \dots, x_n) = \prod_{1 \leq i \leq n} \left( \sum_{1 \leq j \leq n} A(i, j) x_j \right),$$

then  $\frac{\partial^n}{\partial x_1 \dots \partial x_N} p(x_1, \dots, x_n) = Per(A)$  , where  $per(A)$  is the permanent of the matrix  $A$  . Clearly , lower bound  $m \geq \frac{n!}{[\frac{n}{2}]!(n-[\frac{n}{2}])!}$  also holds for multilinear polynomials and even for powers  $(\sum_{1 \leq i \leq n} a_i x_i)^n$  . It is very likely that the actual lower bound is  $2^{n-1}$  and that it does exist somewhere in the geometrical designs literature . In the case of permanents , the formula (7) is essentially the Ryser's formula [4]; and Lemma D.2 says that , in some sense , it is an optimal formula for computing permanents .

Another equivalent formulation of Lemma D.2 is the following statement :

Let a set of complex vectors

$$S = \{X_l = (x_{l,1}, \dots, x_{l,n}) \in C^n : 1 \leq l \leq \frac{(2n-1)!}{(n-1)!n!}\}$$

be a Haar set for the monomials  $M_{r_1, \dots, r_n} : (r_1, \dots, r_n) \in I_{n,n}$ .

I.e. the square matrix  $\{M_{r_1, \dots, r_n}(X_i) : X_i \in S; (r_1, \dots, r_n) \in I_{n,n}\}$  is nonsingular .

If

$$\prod_{1 \leq i \leq n} x_{l,i} = \sum_{1 \leq k \leq m} c_k \left( \sum_{1 \leq i \leq n} Y(k, i) x_i \right)^n$$

for all  $1 \leq l \leq \frac{(2n-1)!}{(n-1)!n!}$  and some complex numbers  $\{c_k; Y(k, i) : 1 \leq k \leq m, 1 \leq i \leq n\}$  then

$$m \geq \frac{n!}{[\frac{n}{2}]!(n-[\frac{n}{2}])!} \approx \frac{2^n}{\sqrt{n}}.$$

■